

Nuclear Energy

New Methods and Tools to Perform Safety Analysis within RISMC

D. Mandelli, C. Smith, C. Rabiti, A. Alfonsi, J. Cogliati, R. Kinoshita

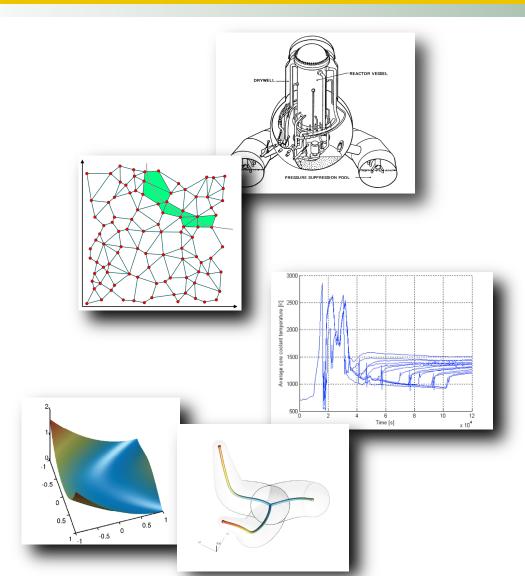
Idaho National Laboratory

ANS Winter Meeting – November 2013



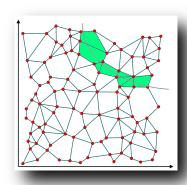


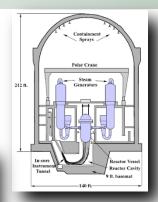
- 1. Modeling (RELAP)
- Generate data
 - Adaptive sampling
 - Reduced order models
 - System emulators
- 3. Analyze time dependent data
 - Clustering
 - Symbolic conversion
- 4. Visualize data
 - Topology based

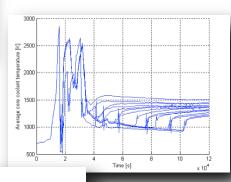


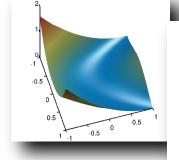


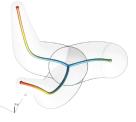
- 1. Modeling (RELAP)
- 2. Generate data
 - Adaptive sampling
 - Reduced order models
 - System emulators
- 3. Analyze time dependent data
 - Clustering
 - Symbolic conversion
- 4. Visualize data
 - Topology based













Data Generation

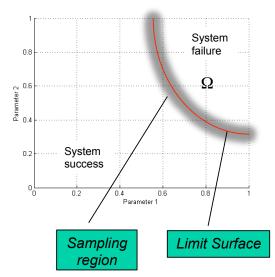
PRA applications

- Determine:
$$p_{failure} = \int_{\Omega} p \, df(\omega) \, d\omega$$

Strategies:

Calculate the integral directly

 Sample timing of events
 Run a single simulation
 Repeat 1. and 2. N times



- Dynamic Event Trees (DETs):
 Branch Scheduler
 System Simulator

Branching occurs when particular conditions have been reached

• Value of specific variables
• Specific time instants
• Plant status

- 2. Evaluate only boundaries of Ω
 - Estimate boundaries
 - Concentrate samples around such boundaries



Data Generation

For large systems, several problems arise if MC or DET are used:

- The set of uncertain parameters is very large
- The computational costs are very high
- Many regions of the input space are not of interest

The space of the possible solutions can be sampled only very sparsely

This precludes the ability to fully analyze the impact of uncertainties on the system dynamics

Understanding of a system depends heavily on where we query

The scope of adaptive sampling is to identify the:

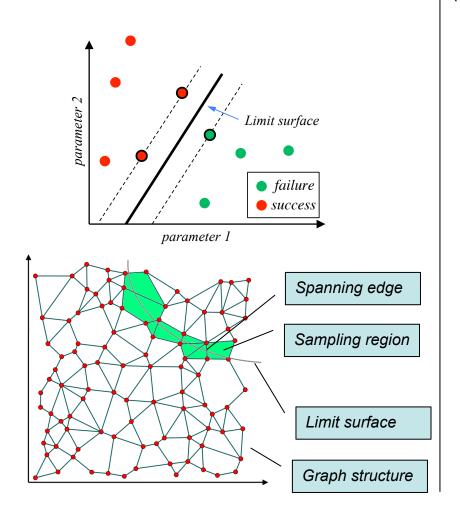
- Set of relevant parameters
- Regions that are of interest for the user

Performed by iteratively guiding the choice of the next sample by analyzing the previous sampling history

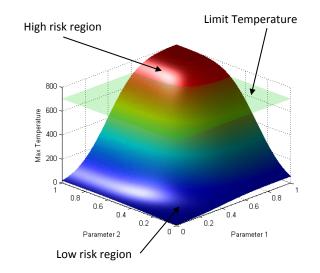


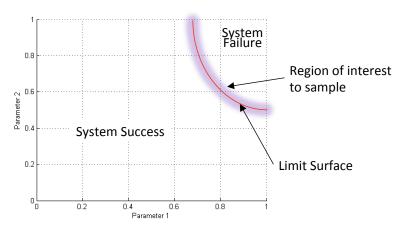
Adaptive Sampling

1- **Data Driven**: Geometric determination of the limit surface



2- **Model Driven**: Prediction of system outcome (e.g., T_{MAX}): surrogate model

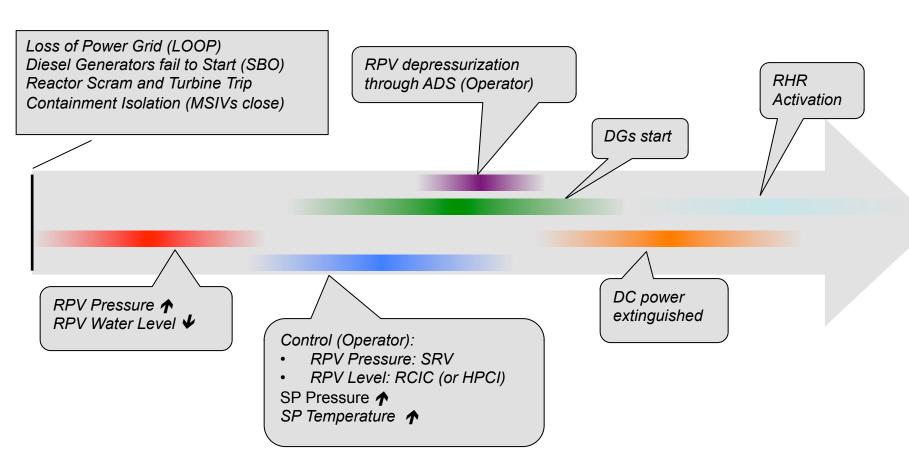






Example of Application: BWR SBO

■ Sequence of events

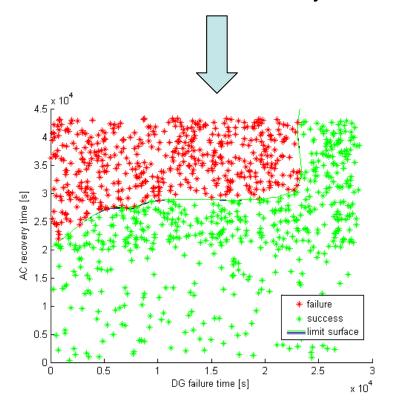


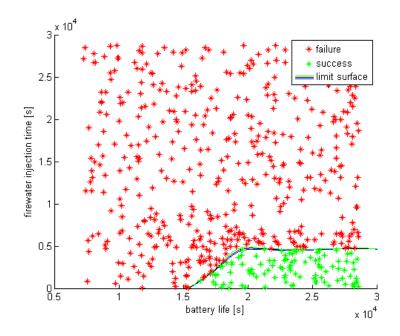


Example of Application: BWR SBO

Fire water injection time vs. battery life

DG failure time vs. AC recovery time



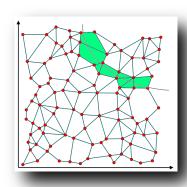


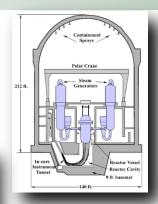
Monte-Carlo sampling: 600 samples

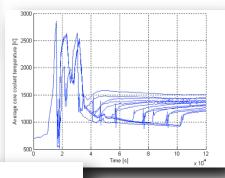
Adaptive sampling: less than 60

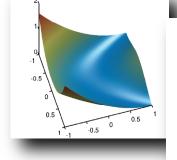


- 1. Modeling (RELAP)
- Generate data
 - Adaptive sampling
 - Reduced order models
 - System emulators
- 3. Analyze time dependent data
 - Clustering
 - Symbolic conversion
- 4. Visualize data
 - Topology based







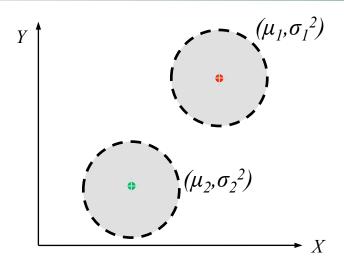


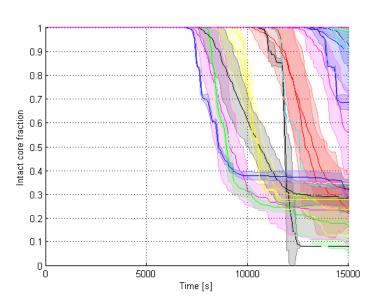




Analyze Time Dependent Data

- Scope: Analyze patterns
- Type of data: thousands of time dependent transients
- Consider the complete time history and not only the end result
- Approach: cluster data into groups
 - Define metric
 - Input clustering level
- Algorithms:
 - Data-centric: K-Means
 - Model-Centric: Density gradient based (Mean-Shift)



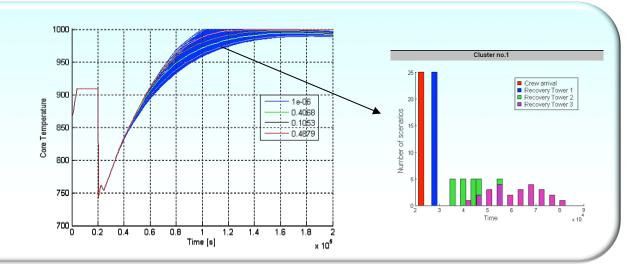




Analyze Time Dependent Data

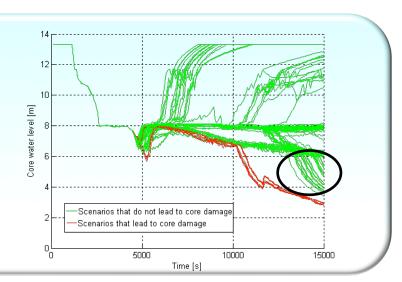
Identify correlations between system dynamics and timing of events

Identify distribution of events for each cluster of scenarios



Evaluate "Near Misses" or scenarios that did not lead to CD because mission time ended before reaching CD

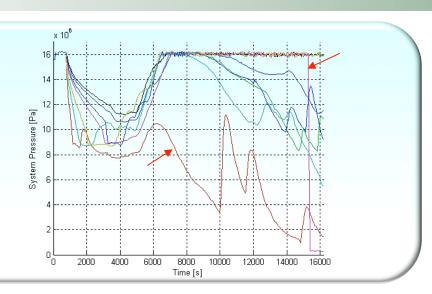
Identify clusters containing scenarios that lead to both system failure and system success



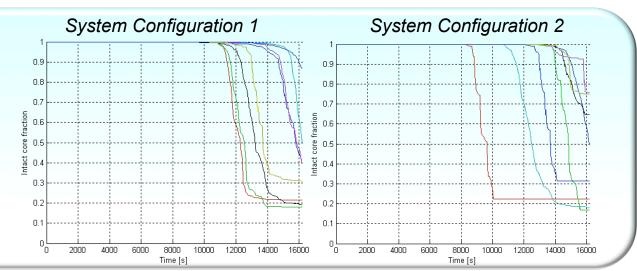


Analyze Time Dependent Data

Identify outliers: "bogus" simulations whose dynamics are different from any other simulation (e.g., out of validity bounds for simulator parameters)

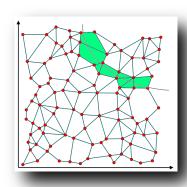


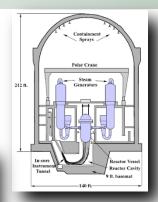
Evaluate system dynamics differences between different sets of analyses (System Design); e.g., different set of system recovery strategies

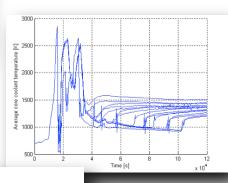


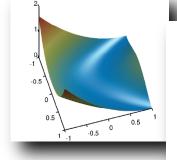


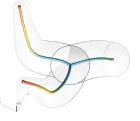
- 1. Modeling (RELAP)
- 2. Generate data
 - Adaptive sampling
 - Reduced order models
 - System emulators
- 3. Analyze time dependent data
 - Clustering
 - Symbolic conversion
- 4. Visualize data
 - Topology based













Data Visualization

INL internally funded project with Scientific Computing and Imaging Institute (University of Utah)

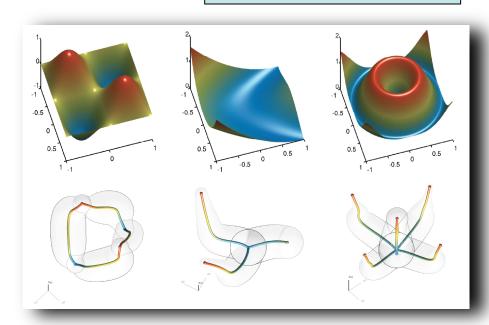
Objective: Develop a software tool to visualize high dimensional data as: $system\ outcome = f(uncertain\ parameters)$

Max clad temperature Max containment pressure

Timing of events HPI water flow rate Initial power

Analysis:

- Exploiting the topological and geometric properties of the domain (Morse-Smale complex)
- Building statistical models based on its topological segmentations
- Providing interactive visual interfaces to facilitate such explorations.

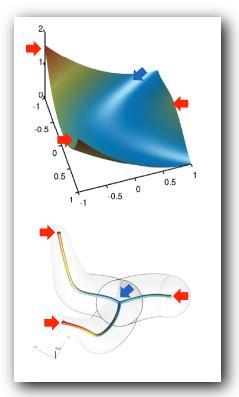


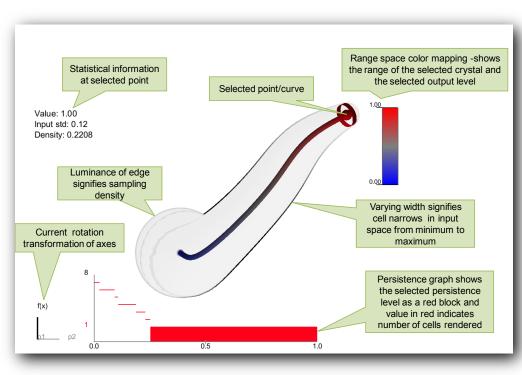
Data Visualization

Graphic overview of HDViz to visualize:

 $system \ outcome = f(input \ parameters)$

■ Visualize the topological structure of *f* through the connection between its min(s) and max(es)



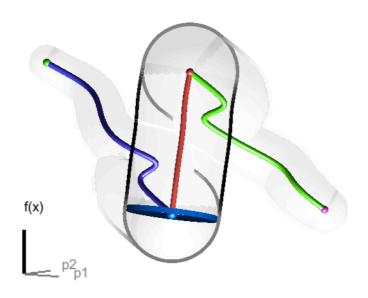


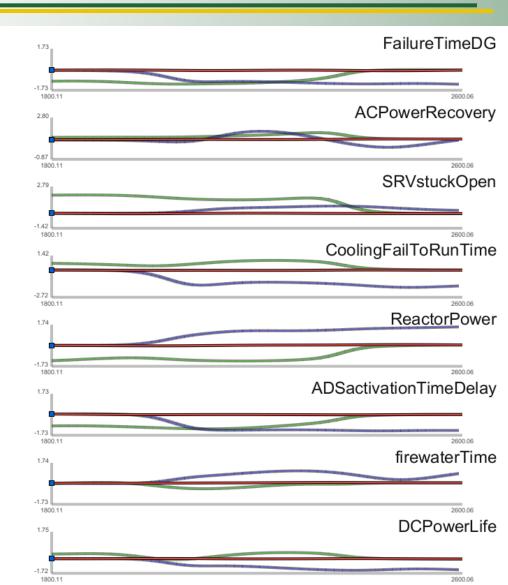


Data Visualization: BWR SBO

Example:

- 20,000 simulation runs
- Outcome: Max clad temperature
- 8 uncertain parameters
- 2 maxima and 2 minima







Simulation Based PRA: Outline

- 1. Modeling (RELAP, MELCOR, MAACS)
- Generate data
 - Adaptive sampling
 - Reduced order models
 - System emulators
- 3. Analyze time dependent data
 - Clustering
 - Symbolic conversion
- 4. Visualize data
 - Topology based

